

**Computing Interval  
Estimates for Components  
of Statistical Information  
with Respect to Judgements  
on Probability Density  
Functions**

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# Imprecise Prevision Theory (IPT)

Starting Points:  
Fundamental Publications  
[1], [2]

- [1] **Walley P.**, *Statistical reasoning with imprecise probabilities*, Chapman and Hall, New York, (1991);
- [2] **Kuznetsov V.**, *Interval statistical models*, Radio and Sviaz, Moscow, (1991) (in Russian).

# Traditional Problem Formulation in the Framework of IPT

Constraints:

$$\rho(x) \geq 0, \quad \int_0^T \rho(x) dx = 1, \quad \text{and}$$
$$\underline{a}_i \leq \int_0^T f_i(x) \rho(x) dx \leq \overline{a}_i, \quad i = 1, 2, \dots, n. \quad (1)$$

It is necessary to find:

$$\inf_{\rho(x)} \int_0^T g(x) \rho(x) dx \quad \text{as well as}$$
$$\sup_{\rho(x)} \int_0^T g(x) \rho(x) dx \quad (2)$$

subject to constraints (1).

# Dual for the Initial Problem Statement

Let us find:

$$\underline{M}(g) = \sup_{c_0, c_i, d_i} \left( c_0 + \sum_{i=1}^n (c_i \underline{a}_i - d_i \bar{a}_i) \right) \quad (3)$$

subject to  $c_0 \in R$ ,  $c_i, d_i \in R_+$  and for any  $x \geq 0$ ,

$$c_0 + \sum_{i=1}^n (c_i - d_i) f_i(x) \leq g(x). \quad i=1, 2, \dots, n: \quad (4)$$

And

$$\bar{M}(g) = \inf_{c_0, c_i, d_i} \left( c_0 + \sum_{i=1}^n (c_i \bar{a}_i - d_i \underline{a}_i) \right) \quad (5)$$

subject to  $c_0 \in R$ ,  $c_i, d_i \in R_+$  and for any  $x \geq 0$ ,

$$c_0 + \sum_{i=1}^n (c_i - d_i) f_i(x) \geq g(x). \quad i=1, 2, \dots, n: \quad (6)$$

# Important Conclusion Concerning Optimal Solutions

(L. Utkin and I. Kozine, [3])

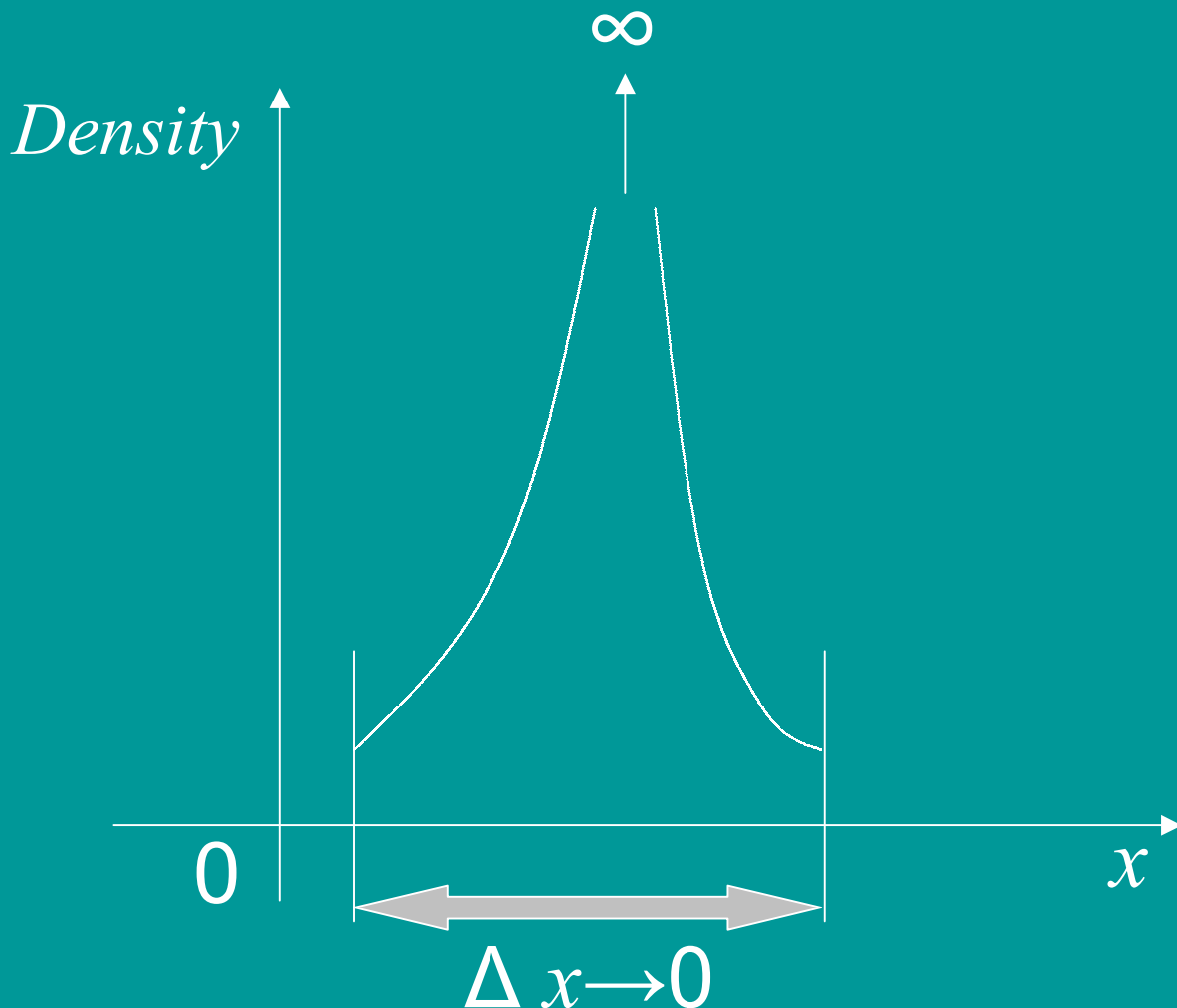
[3] Utkin L. and Kozine I.

*Different faces of the natural extension.* In: Proceedings of the Second International Symposium on Imprecise Probabilities and Their Applications, ISIPTA '01, 2001, pp. 316-323.

Optimal solutions belong to a family of **DEGENERATE** distributions (such probability densities are composed of  $\delta$ -functions)

# Distribution of Probabilistic Masses

Masses are concentrated in the fixed points:

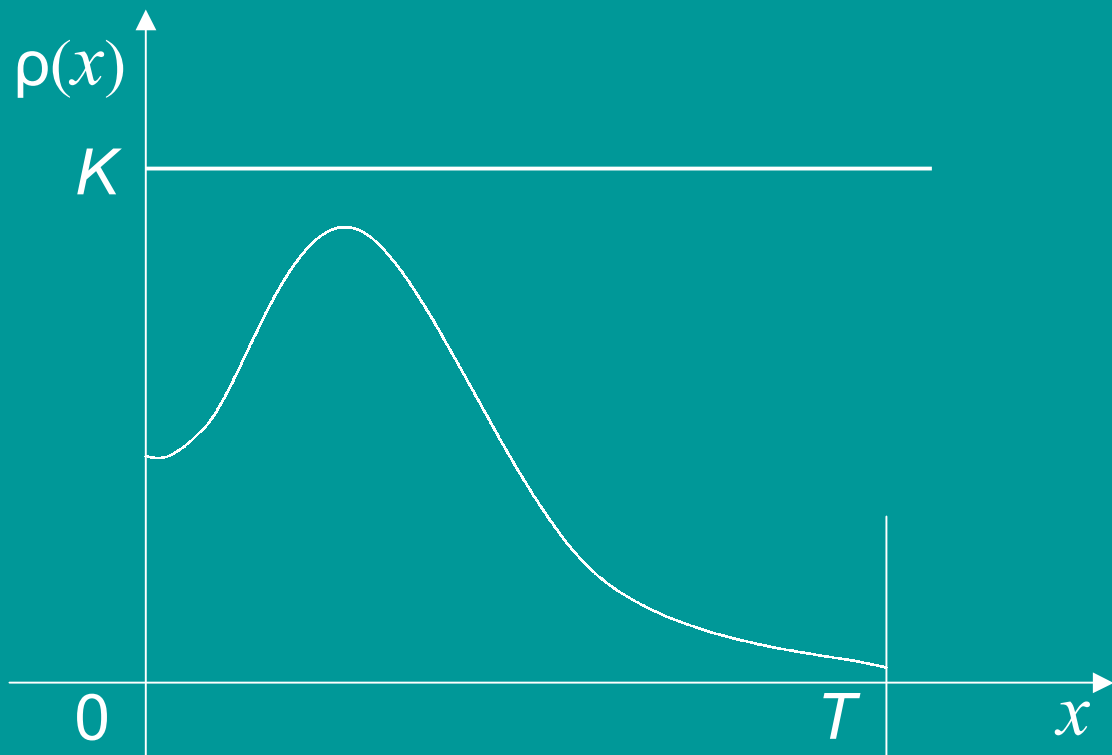


# Use of Additional Judgements

Additional judgement can be reflected by inequality:

$$\rho(x) \leq K = \text{const}, \quad (7)$$

where  $K \in R_+$  is such that  $KT \geq 1$ .



# Main Goal (Theorem)

*If there is no any finite interval  $x \in [\alpha, \beta] \subseteq [0, T]$ ,  $\alpha < \beta$ , for which function  $g(x)$  can be represented in the form*

$$g(x) = h_0 + \sum_{i=1}^n h_i f_i(x), \quad (8)$$

*where  $h_0, h_1, \dots, h_n \in R$ , then function  $\rho(x)$  providing solution of optimization problem mentioned above, belongs to class of step-functions with minimum value equal to 0 and maximum value equal to  $K$ .*

# Some Comments

To provide (8) the system

$$h_0 + \sum_{i=1}^n h_i f_i(x) = g(x),$$

$$\sum_{i=1}^n h_i f_i'(x) = g'(x),$$

$$\sum_{i=1}^n h_i f_i''(x) = g''(x),$$

.....

$$\sum_{i=1}^n h_i f_i^{(n)}(x) = g^{(n)}(x)$$

must have at least one solution which is independent on  $x$  in some interval  $x \in [\alpha, \beta]$ .

# Applying Methodology of the Calculus of Variations

The inequalities

$$0 \leq \rho(x) \leq K \quad (9)$$

should be excluded from direct consideration in order to allow operating in the open domain with the values of the function.

The requirement  $\rho(x) \geq 0$  can be replaced by denoting

$$\rho(x) = z^2(x). \quad (10)$$

The requirement  $\rho(x) \leq K$  can be reflected by equality

$$z^2(x) + v^2(x) = K, \quad (11)$$

where  $v(x)$  is newly introduced function.

# Modified Formulation of the Problem

We would like to estimate

$$\inf_{z(x)} \int_0^T g(x) z^2(x) dx$$

and

$$\sup_{z(x)} \int_0^T g(x) z^2(x) dx \quad (12)$$

subject to

$$z^2(x) + v^2(x) = K, \quad (13)$$

$$\int_0^T z^2(x) dx = 1, \quad (14)$$

$$\int_0^T f_i(x) z^2(x) dx \leq \bar{a}_i, \quad (15)$$

$$-\int_0^T f_i(x) z^2(x) dx \leq \underline{a}_i, \quad i=1,2,\dots,n. \quad (16)$$

# Lagrange Approach

$$F(z, v) = g(x)z^2(x)$$



$$F^*(z, v) = g(x)z^2(x) + \lambda^*(x) \cdot (z^2(x) + v^2(x)) + \lambda_0 z^2(x) + \sum_{i=1}^n \lambda_i f_i(x) z^2(x) - \sum_{i=n+1}^{2n} \lambda_i f_i(x) z^2(x)$$

Equations of Euler – Lagrange:

$$\frac{\partial F^*(z, v)}{\partial z} - \frac{d}{dx} \left( \frac{\partial F^*(z, v)}{\partial \dot{z}} \right) = 0; \quad (17)$$

$$\frac{\partial F^*(z, v)}{\partial v} - \frac{d}{dx} \left( \frac{\partial F^*(z, v)}{\partial \dot{v}} \right) = 0.$$

# The Necessary Conditions of Optimality

The equations look here as follows:

$$z(x) \cdot \left( g(x) + \lambda^*(x) + \lambda_0 + \sum_{i=1}^n (\lambda_i - \lambda_{i+n}) f_i(x) \right) = 0; \quad (18)$$

$$\lambda^*(x) v(x) = 0.$$

Let us fix any interval

$$x \in [\alpha, \beta] \subseteq [0, T], \quad \alpha < \beta.$$

**Case 1.**  $z(x) \equiv 0$

inside the interval.

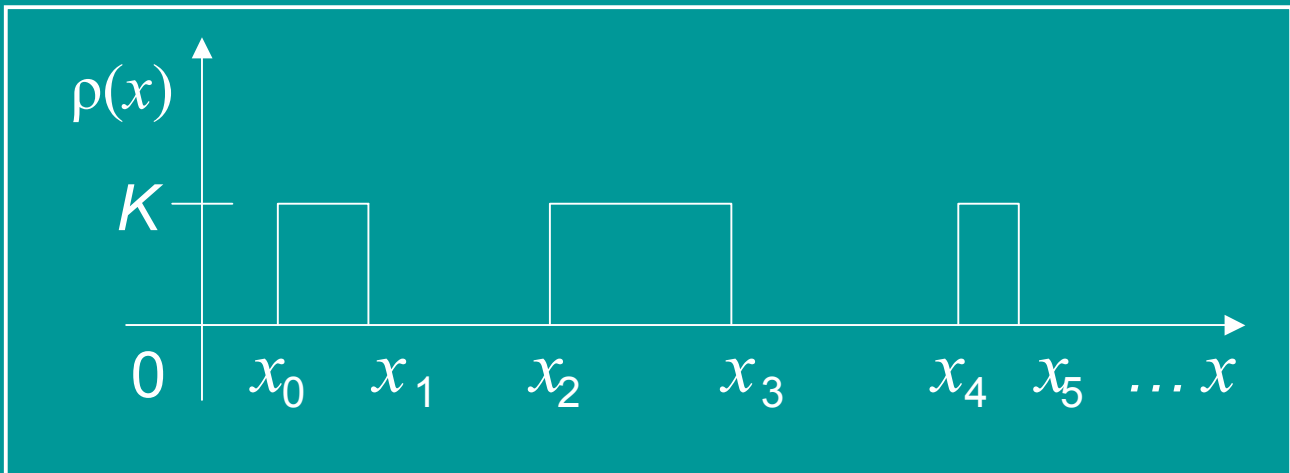
Then  $v(x) = \sqrt{K}$  and  $\lambda^*(x) = 0$ .

**Case 2.**  $z(x) \neq 0$ , so

$$v(x) = 0 \quad \text{and} \quad z(x) = \sqrt{K}.$$

# Practical Implementation

Optimal probability density:



Denote:

$$G(x_j, x_{j+1}) = \int_{x_j}^{x_{j+1}} g(x) dx, \quad (19)$$

$$\Phi_i(x_j, x_{j+1}) = \int_{x_j}^{x_{j+1}} f_i(x) dx, \quad i = 1, 2, \dots, n. \quad (20)$$

# Reformulation of the Problem Statement

We would like to estimate

$$\min_{x_0, x_1, \dots} \left\{ K \cdot \sum_{j=0}^m G(x_{2j}, x_{2j+1}) \right\}, \quad (21)$$

$$\max_{x_0, x_1, \dots} \left\{ K \cdot \sum_{j=0}^m G(x_{2j}, x_{2j+1}) \right\} \quad (22)$$

subject to

$$K \cdot \sum_{j=0}^m (x_{2j+1} - x_{2j}) = 1; \quad (23)$$

$$\underline{a}_i \leq K \cdot \sum_{j=0}^m \Phi_i(x_{2j}, x_{2j+1}) \leq \bar{a}_i, \quad i = 1, 2, \dots, n. \quad (24)$$

# Example 1

The information concerning a continuous random variable  $X$  is  $\rho(x) \leq \Psi(x) = K \cdot I_{[0,T]}(x) < \infty$ , where  $K, T$  are fixed positive numbers. What are the **bounds** for the **expectation**  $M(X)$ ?

\* \* \*

Let us choose  $m=0$ .

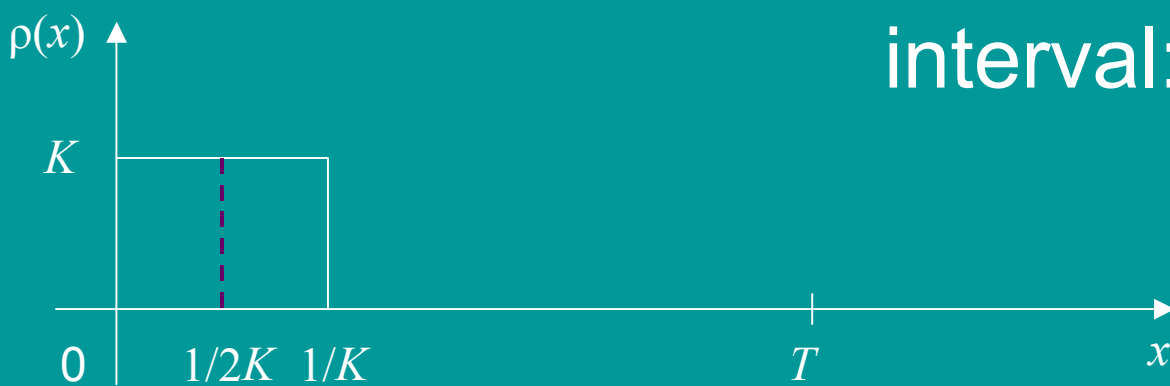
Objective function:

$$J = \int_0^T x \rho(x) dx = K \cdot \int_{x_0}^{x_1} x dx = K \cdot \frac{x_1^2 - x_0^2}{2}.$$

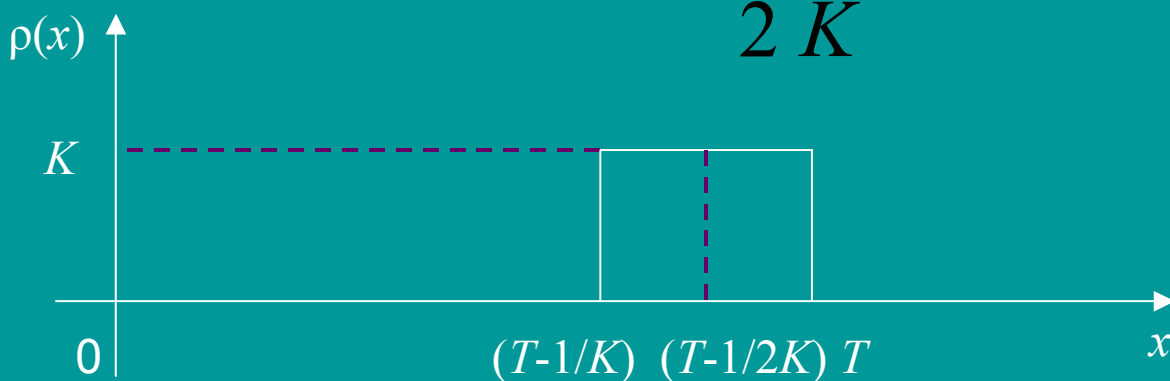
# Solution of Optimization Problem

Lower and upper bounds of  $J$

interval:



$$\min J = \frac{1}{2K}.$$



$$\max J = (T - 1/K) + \frac{1}{2K} = T - \frac{1}{2K}.$$

## Example 2

We add the constraint:

$$\int_0^T f_1(x) \rho(x) dx = b,$$

where  $f_1(x) = I_{[\underline{a}, \bar{a}]}(x)$  is the indicator function.

Here also any finite interval of  $x$  values for which

$$g(x) = x = c_0 + c_1 I_{[\underline{a}, \bar{a}]}(x)$$

cannot be found, so the theorem can be applied.

Further analysis shows, that  $m=1$  is the best choice for such situation.

## Example 2 (Continuation 1)

To provide  $J \rightarrow \min$   
we have to set:

(i) if  $\underline{a} \geq \frac{1-b}{K}$  :

$$\rho(x) = \begin{cases} K & \text{for } 0 \leq x \leq \frac{1-b}{K}; \\ 0 & \text{for } \frac{1-b}{K} < x < \underline{a}; \\ K & \text{for } \underline{a} \leq x \leq \underline{a} + \frac{b}{K}; \\ 0 & \text{for } \underline{a} + \frac{b}{K} < x \leq T; \end{cases}$$

## Example 2 (Continuation 2)

(ii) if  $\underline{a} < \frac{1-b}{K}$ :

$$\rho(x) = \begin{cases} K & \text{for } 0 \leq x \leq \underline{a} + \frac{b}{K}; \\ 0 & \text{for } \underline{a} + \frac{b}{K} < x < \bar{a}; \\ K & \text{for } \bar{a} \leq x \leq \bar{a} + \frac{1 - \underline{a}K - b}{K}; \\ 0 & \text{for } \bar{a} + \frac{1 - \underline{a}K - b}{K} < x \leq T. \end{cases}$$

As the result

$$\min J = \frac{1}{2K} + \frac{b(\underline{a}K - (1-b))}{K} \quad \text{or}$$

$$\min J = \frac{1}{2K} + \frac{\bar{a}K - (\underline{a}K + b) \cdot (1 + (\bar{a} - \underline{a})K - b)}{K}.$$

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